Tweet Emotion Dynamics: Emotion Word Usage in Tweets from US and Canada

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We introduce:

- TUSC: ~45 million geo-located Tweets from US,
 Canada
- Tweet Emotion Dynamics (TED): a set of metrics to quantify emotion word usage in tweets across time

Why Track Emotion Word Usage

 Over Time: Are we tweeting more positive words, negative words, high arousal words, etc. over time?
 How has the COVID-19 pandemic impacted tweets?

When did we use the most number of words conveying a lack of control and uncertainty?

How were individual cities impacted? etc.

 Geographically: How are Canada and US different in terms of emotion word usage?

The TUSC Dataset (after appropriate preprocessing of tweets)

- TUSC-Country: tweets from US and Canada (CA)
- ~103K tweets per year, 2015–2018
- ~380K tweets per year, 2019–2021
- TUSC-City: tweets from 46 US–CA cities, Apr 2020–Dec 2021
- ~26M tweets per year

Valence

Fun fact: CA tweets, on average, use two more tokens per tweet than US tweets (less informal, slang, etc. (Snefjella et al. 2018))

Arousal

Utterance (e.g., Tweet) Emotion Dynamics

- Emotion Dynamics: A framework from Psychology for measuring changes in one's emotions over time.
- Utterance Emotion Dynamics (UED) Hipson and Mohammad (2020): A computational framework for changes in emotions associated with *utterances* over time.
- Tweet Emotion Dynamics (TED): Our use of UED on tweets.

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Set Up

- Create temporally ordered stream of words by speaker
- Apply a rolling window averaging word–emotion association scores
- This sequence of window emotion scores is the **emotion arc** of their utterances

Emotion dimensions

- Valence (V): positive—negative pleasure—displeasure
- Arousal (A): active—sluggish
- Dominance (D): powerful—weak in control—no control

NRC VAD Lexicon

- Scores between 0 and 1 for the V, A, D for ~20,000 English words
- Removed 'neutralish' words (0.33--0.66)
- Removed frequent ambiguous words (trump, may, will)

Notable UED Metrics

- Home Base: Range of emotion scores one standard deviation away from the mean on each side. Most probable emotion space occupied by speaker.
- Variability: Standard deviation of emotion means.
- Rise Rate: Rate at which speaker reaches peak emotional intensity (emotional reactivity).
- Recovery Rate: Rate at which speaker returns to home base (emotion regulation).

EXPERIMENTS

1. Average Emotion Scores of Words in Tweets (How emotional are our tweets?)

Dominance

Year	Canada	USA	Canada	USA	Canada	USA
2015 TUSC	0.675	0.644	0.461	0.465	0.564	0.535
2016 Country	0.672	0.644	0.464	0.469	0.567	0.543
2017	0.666	0.639	0.465	0.472	0.568	0.546
2018	0.669	0.640	0.463	0.472	0.576	0.553
2019	0.668	0.638	0.461	0.476	0.577	0.553
20152019	0.670	0.641	0.463	0.471	0.570	0.546
2020	0.662	0.634	0.458	0.474	0.574	0.555
2021	0.669	0.644	0.457	0.474	0.578	0.560
2020 TUSC	0.651	0.634	0.469	0.479	0.571	0.558
2021 City	0.658	0.645	0.470	0.479	0.576	0.563

CA (vs US): higher V, lower A, higher D
Yearly:

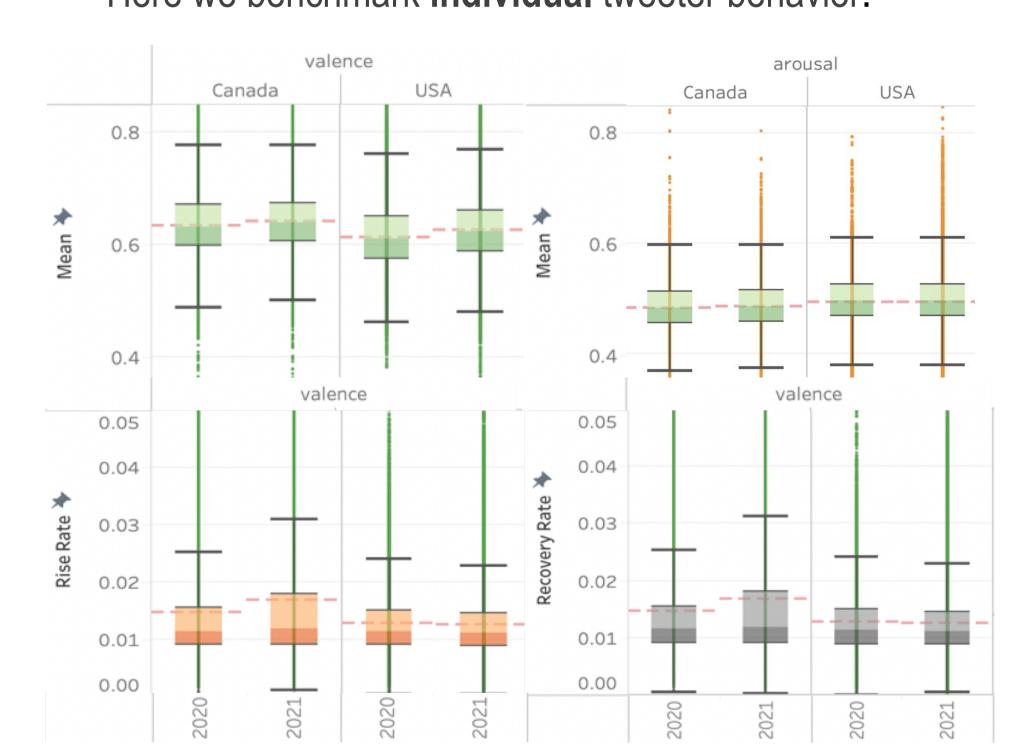
- V: Lowest in 2020 (June 2020)
- A: US scores increased with time
- D: Steady increase with time

Pandemic impact: lower V, lower D

TUSC-City, Country: similar trends

3. UED Metrics

Exps 1,2: aggregate emotion word usage (city, country) Here we benchmark **individual** tweeter behavior.



2. Proportion of Tweets with Emotional Terms (How often are we tweeting emotional terms?)

Find proportion of tweets with at least one high V/A/D word

Divided lexicons into low (score <= 0.33) and high (score >= 0.67) sub-lexicons.

	Low Va	Low Valence		High Valence		Low Arousal		High Arousal	
Year	Canada	USA	Canada	USA		Canada	USA	Canada	USA
2015	36.3	38.4	80.2	75.4		55.8	51.1	40.9	38.1
2016	37.6	39.6	80.8	76.7		56.4	52.0	42.1	40.0
2017	40.8	43.2	82.9	79.0		59.4	55.6	45.0	43.5
2018	42.7	45.7	83.2	80.2		61.4	57.8	47.1	46.0
2019	42.9	45.3	82.7	79.4		60.8	56.7	46.5	46.1
20152019	40.0	42.4	82.0	78.1		58.8	54.6	44.3	42.7
2020	43.8	46.1	82.3	79.2		60.8	57.0	45.7	45.9
2021	42.6	44.5	82.5	79.7		61.1	57.2	45.6	46.1

High-V (vs Low-V): ~ 100% more Low-A (vs High-A): ~ 40% more High-D (vs Low-D): ~33% more

CA (vs US): more high-V, fewer low-V more low-A, high-D

2020: higher low-V, fewer high-V

We tweeted fewer positive AND more negative words

- Overall: Metrics show Gaussian distribution
- Mean: Similar VAD trends across CA–US as in 1, 2
- Rise, Recovery Rates: Larger third quartile for CA
 V and D: rise, recovery rates for positive deviations
 - from home base lower than for negative deviations
 Trend is reversed for A
- City-level (figure not shown here):
- Highest V: London, Ottawa, Halifax, Victoria (CA)
- Lowest V: Detroit, Houston, LA, Philadelphia (US)

Ethical Considerations

Before you start:

- 1. Ethics Sheet for Automatic Emotion Recognition (Mohammad, 2022)
- 2. Practical and Ethical Considerations in the Use of Emotion Lexicons (Mohammad, 2020)

Where Do We Go From Here?

- TED framework can be applied to any set of temporally-ordered utterances
- Applications in domains such as digital humanities, social sciences, psychology.
- Collaboration with UNC Carolina Affective Science Lab:
 - Can TED inform us about the mental and physical health of populations?











